

Iterative Coding for Network Coding

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March 26, 2008

Outline

- 1 Information is not oranges
- 2 Practical network coding
- 3 Noisy network coding
- 4 Construction and decoding algorithm
- 5 Achieving capacity

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Information is not oranges

Practical network coding

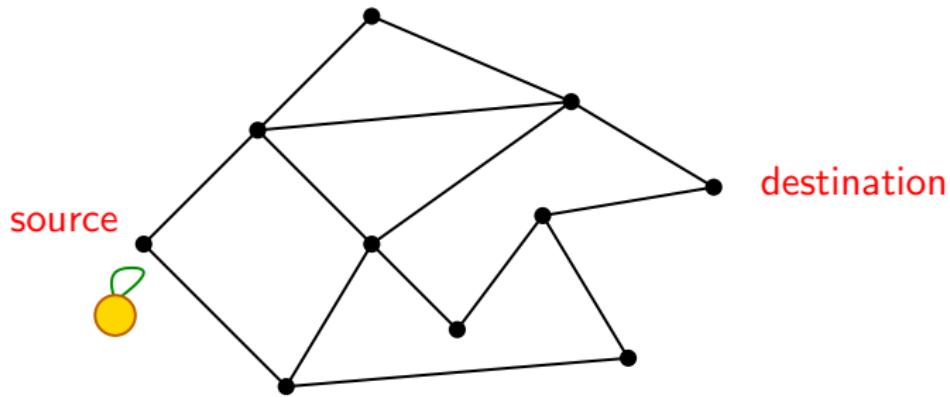
Noisy network coding

Construction and decoding algorithm

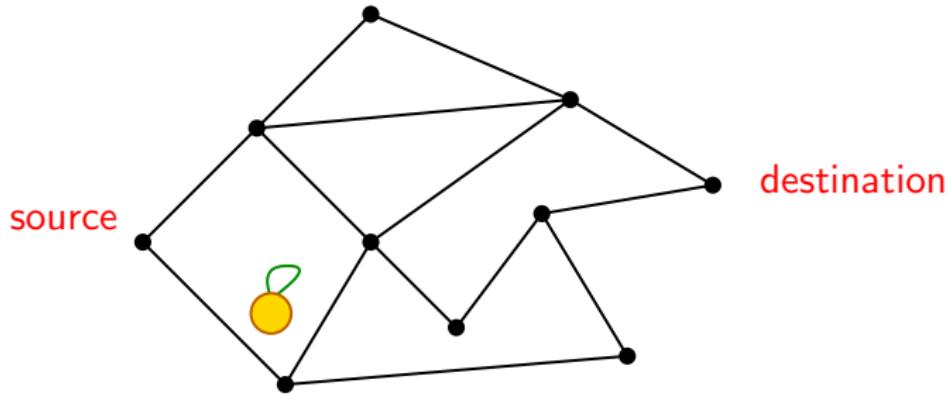
Achieving capacity

Information is not oranges

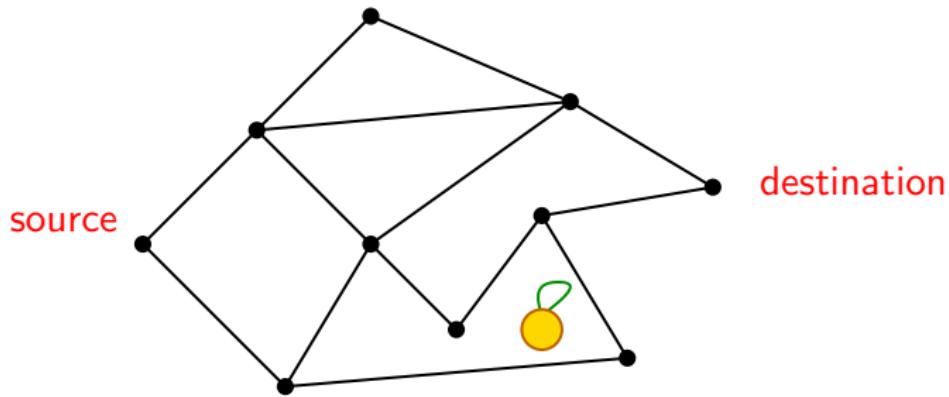
Current approach to communication networks



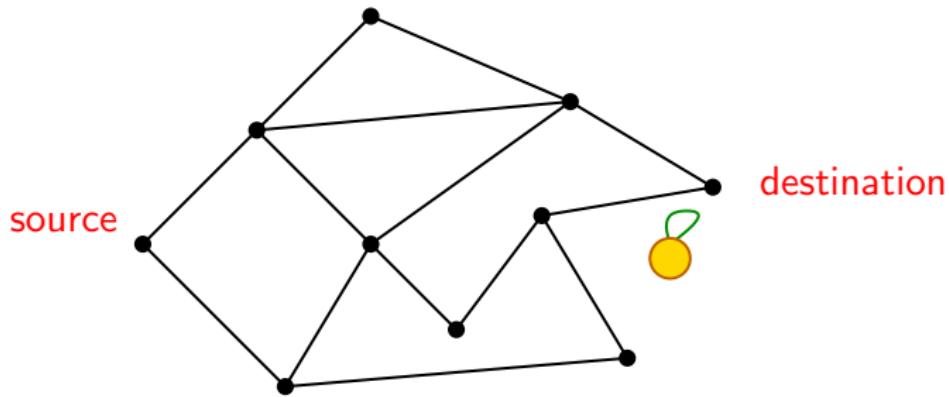
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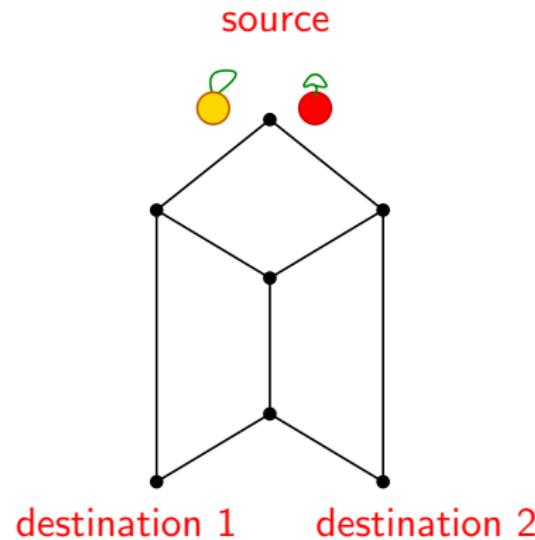
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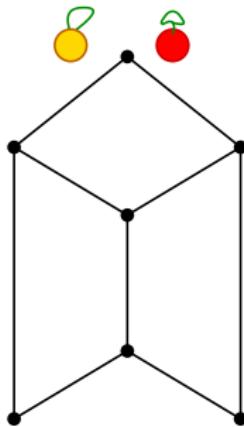


What are we losing? The butterfly example

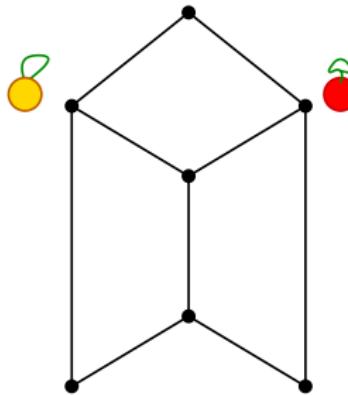


[Ahlswede,Cai,Li,Yeung, 2000]

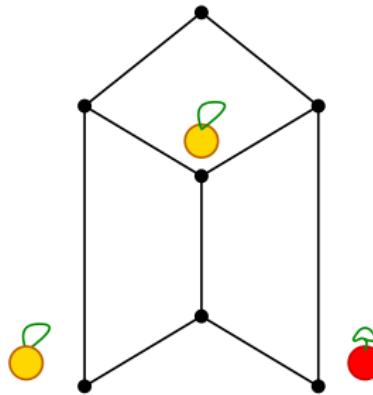
The butterfly example: Routing



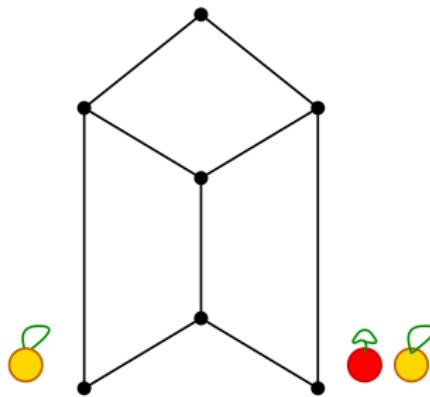
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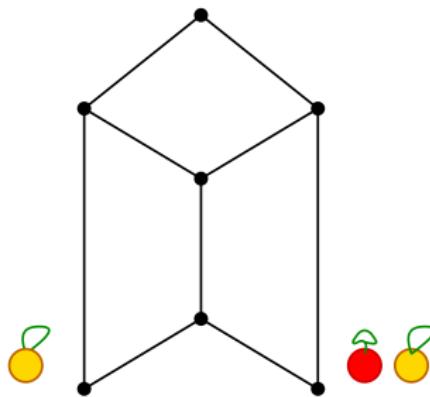
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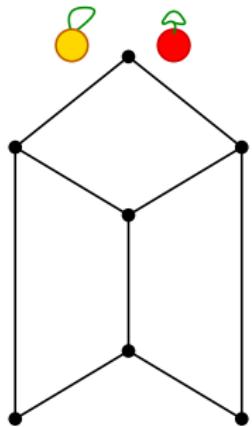


The butterfly example: Routing

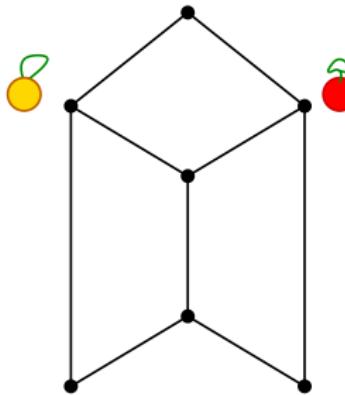


1.5 bits per cycle

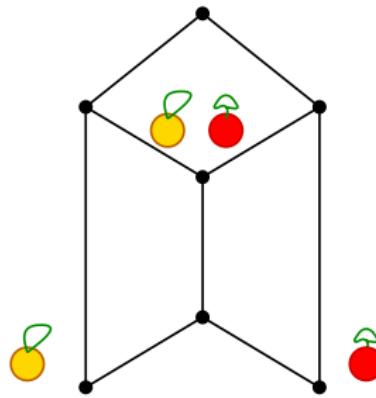
The butterfly example: Network Coding



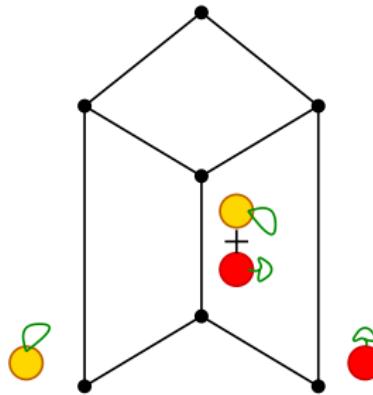
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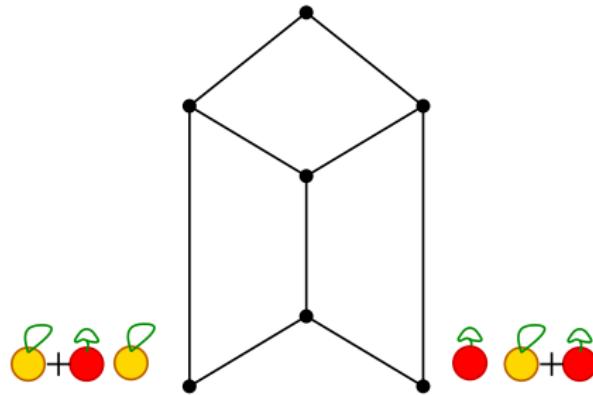
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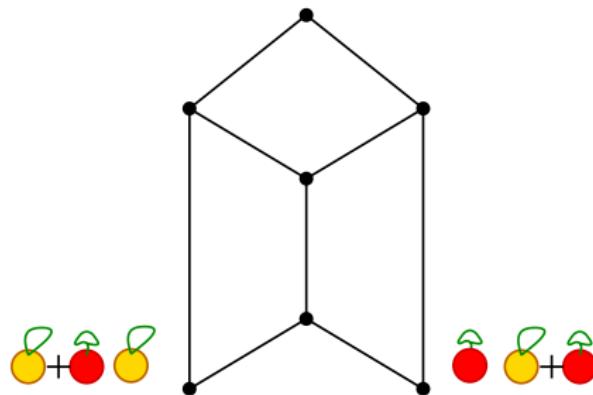
The butterfly example: Network Coding



The butterfly example: Network Coding



The butterfly example: Network Coding



2 bits per cycle

Information is not oranges

Practical network coding

Noisy network coding

Construction and decoding algorithm

Achieving capacity

Practical network coding

Problem

No one knows the network structure.

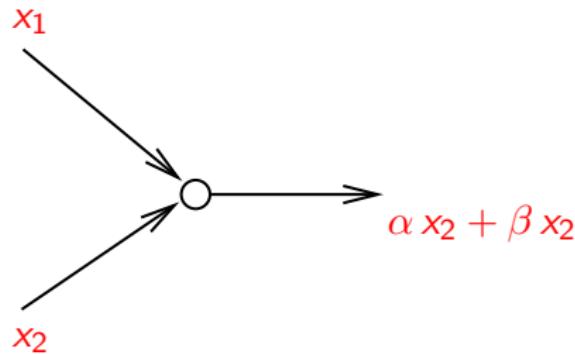
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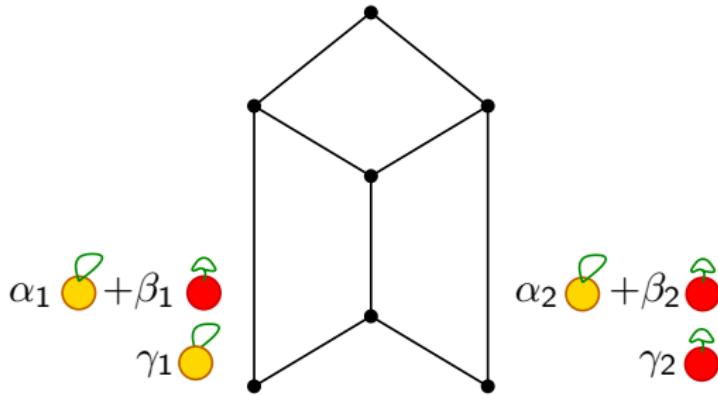
Idea



Forward random combinations

[Chou, Wu, Jain/ Ho, Kötter, Medard, Karger, Effros 2003]

How can that possibly work?



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Wait a minute...

How am I supposed to figure out α, β, γ ?

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Wait a minute...

How am I supposed to figure out α, β, γ ?

Idea: Header

Input:

 $= [1 \quad 0 | \dots \dots \quad x_1 \dots \dots]$

 $= [0 \quad 1 | \dots \dots \quad x_2 \dots \dots]$

Idea: Header

Output:

$$? \text{ } \textcolor{red}{\text{apple}} + ? \text{ } \textcolor{yellow}{\text{orange}} = [\alpha \text{ } \gamma | \cdots \alpha x_1 + \gamma x_2 \cdots]$$

$$? \text{ } \textcolor{red}{\text{apple}} + ? \text{ } \textcolor{yellow}{\text{orange}} = [\delta \text{ } \beta | \cdots \delta x_1 + \beta x_2 \cdots]$$

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Noisy network coding

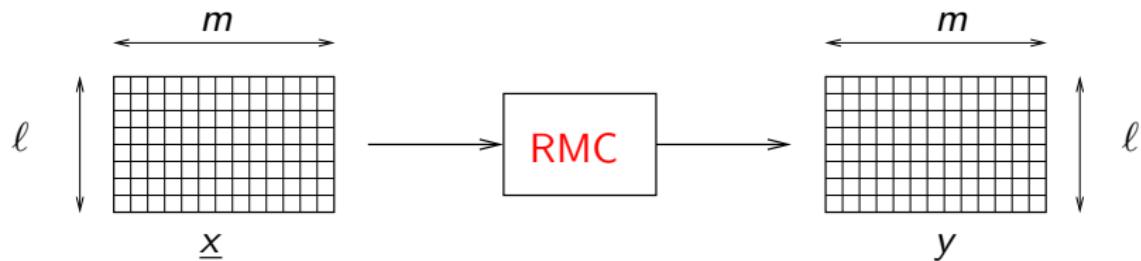
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Noisy network coding

Rank Metric Channel (Symmetric Network Channel)

(Gabidulin, 1985)



$$\underline{y} = \underline{x} \oplus \underline{z},$$

\underline{z} uniformly random with $\text{rank}(\underline{z}) = \ell\omega$

Asymptotics, Rate, Capacity

$$\ell = N\lambda, m = N(1 - \lambda), N \rightarrow \infty$$

$$R = \frac{\log_2 |\text{Code}|}{m\ell},$$

$$C(\lambda, \omega) = \frac{1 - \lambda - \omega + \lambda\omega^2}{1 - \lambda}.$$

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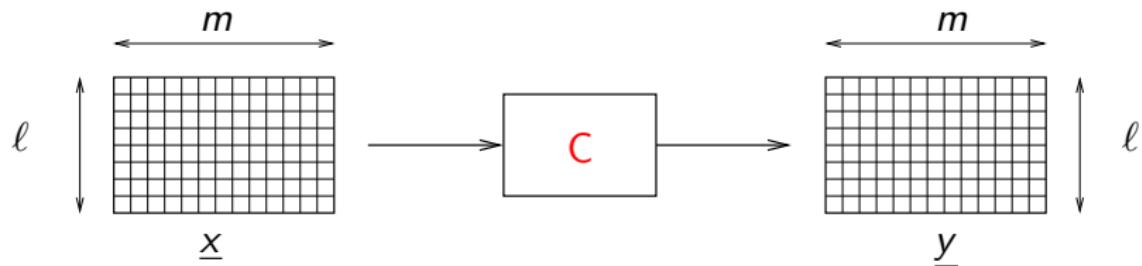
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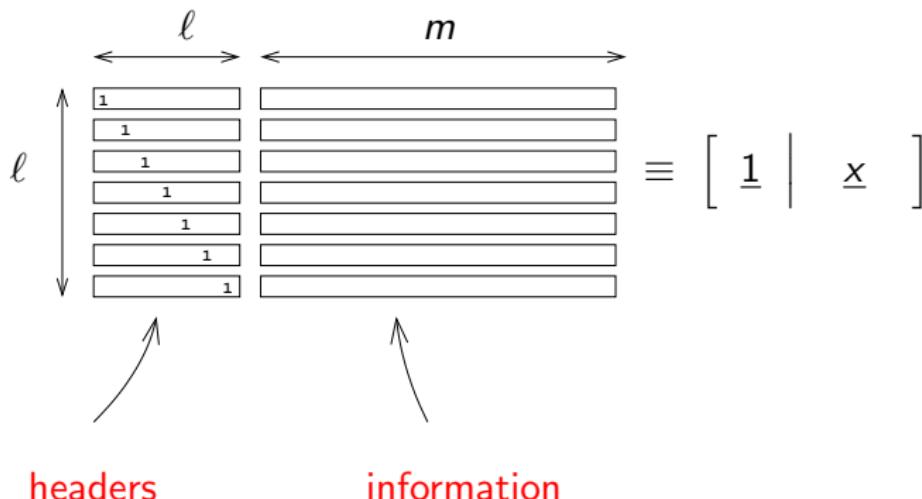
Matrix channels



$$\underline{x} = \{x_{ij}\} \in \mathbb{F}_2^{m \times \ell}$$

Network coding???

(Kötter, Kschischang, 2007)



(Chou, Wu, Jain, 2003)

Reliable network

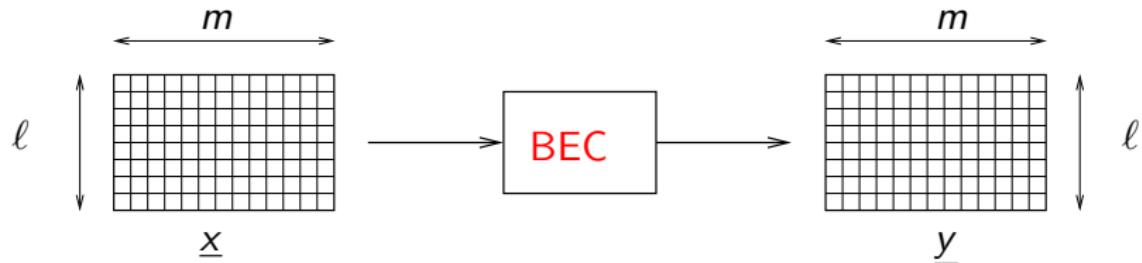


Faulty network

$$\left[\begin{array}{c|c} 1 & \underline{x} \end{array} \right] \xrightarrow{\text{NET}} \boxed{\text{NET}} \xrightarrow{\quad} \left[\begin{array}{c|c} G & G\underline{x} + \underline{z}' \end{array} \right] \xrightarrow{\quad} \boxed{G^{-1}} \xrightarrow{\quad} \left[\begin{array}{c|c} 1 & \underline{x} + G^{-1}\underline{z}' \end{array} \right]$$

Binary Erasure Channel

(Elias, 1954)



$$y_{ij} = \begin{cases} x_{ij} & \text{with prob. } \epsilon, \\ * & \text{with prob. } 1 - \epsilon. \end{cases} \quad \text{independently}$$

Asymptotics, Rate, Capacity

$$\ell = N\lambda, m = N(1 - \lambda), N \rightarrow \infty$$

$$R = \frac{\log_2 |\text{Code}|}{m\ell},$$

$$C(\epsilon) = 1 - \epsilon.$$

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Practical network coding

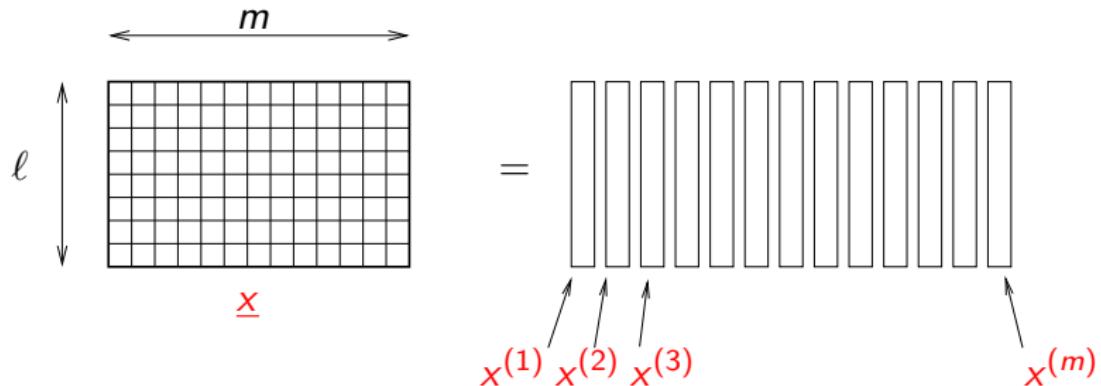
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Construction and decoding algorithm

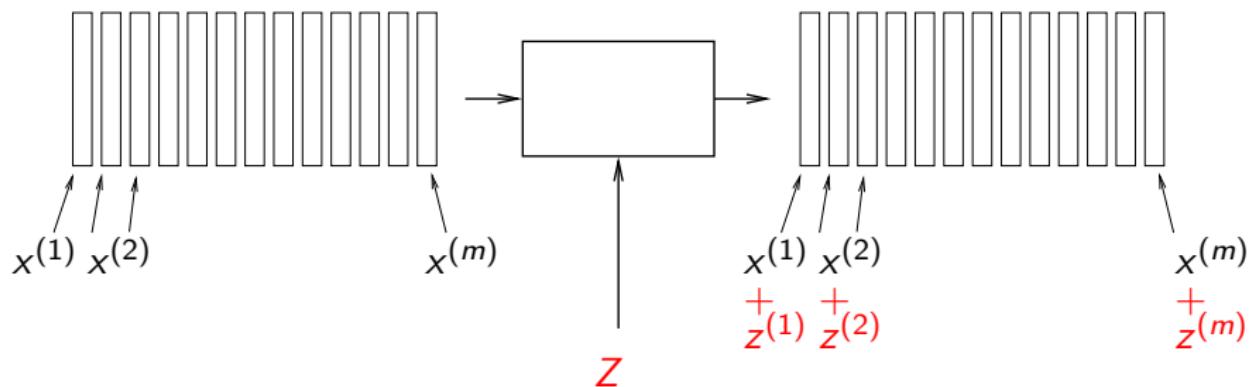
Achieving capacity

Construction and decoding algorithm

Equivalent descriptor of the channel

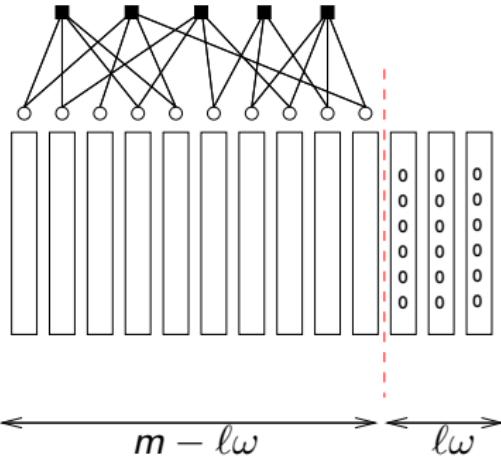


Equivalent descriptor of the channel



$Z \equiv$ uniformly random subspace $\subseteq \mathbb{F}_2^\ell$
 $z^{(1)}, \dots, z^{(m)} \in Z$

Code

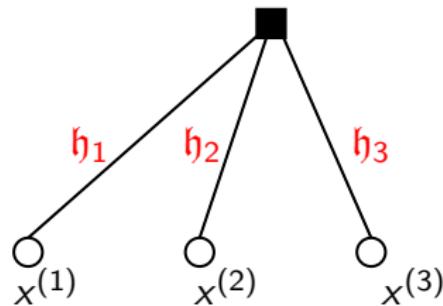


$m - \ell\omega$ 'symbols' → LDPC

$\ell\omega$ 'symbols' → learn the error space Z

LDPC: Check nodes

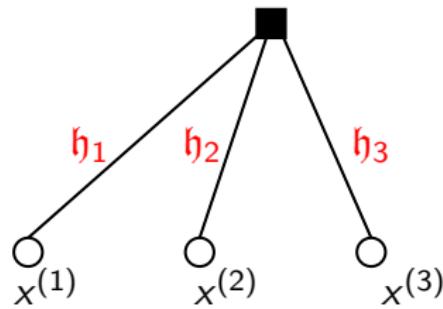
$$\mathfrak{h}_1 x^{(1)} + \mathfrak{h}_2 x^{(2)} + \mathfrak{h}_3 x^{(3)} = 0$$



'Edge labels' $\mathfrak{h}_i \in \mathbb{F}_2^{\ell \times \ell}$: $\ell \times \ell$ matrices

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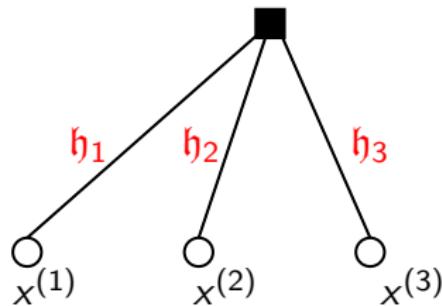
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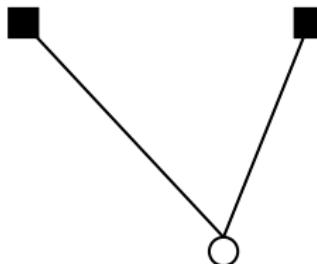
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LDPC: Variable nodes



$$x^{(i)} \in \{z^{(i)} + Z\}$$

Degree = 2 !

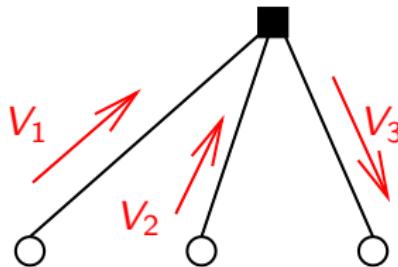
Message passing decoder

Messages \rightarrow Affine subspaces $V_{i \rightarrow a} \subseteq \mathbb{F}_2^\ell$

Operations \rightarrow Subspace intersections/sums

Message passing decoder: Check nodes

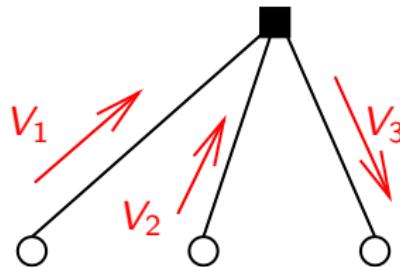
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$$V_3 = \mathfrak{h}_3^{-1}(\mathfrak{h}_1 V_1 + \mathfrak{h}_2 V_2)$$

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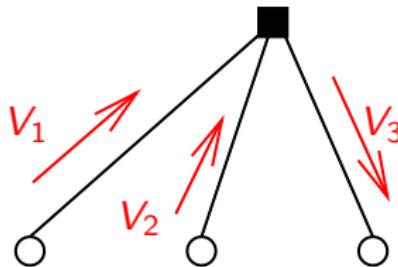
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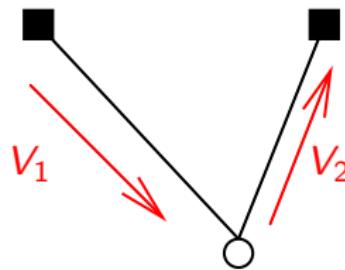
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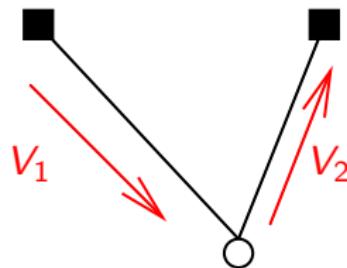
Message passing decoder: Variable nodes



$$x^{(i)} \in \{z^{(i)} + Z\}$$

$$V_2 = V_1 \cap \{z^{(i)} + X\}$$

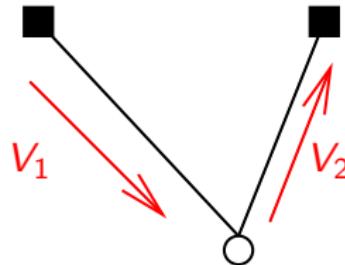
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Achieving capacity

Capacity achieving ensemble

Variable degree:

2

Check degree:

$$P_i^{(k)} = \frac{2k(k-1)}{i(i-1)(i-2)}.$$

Not the soliton!

Capacity achieving ensemble

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Capacity achieving ensemble

Theorem

If $\omega = 1/k$, then the ensemble $(2, P^{(k)})$ has rate equal to the capacity of the rank metric channel and achieves vanishing error probability under message passing decoding.

Further, it achieves vanishing error probability over the erasure channel.

Capacity achieving ensemble

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Encoding complexity $O(m\ell^2)$

Decoding complexity $O(m\ell^2)$

Error probability $\exp[-\Omega(\gamma^{\text{iter}})], \exp(-\Omega(m, \ell))$

Realistic example (from Chou et al.):

$$\ell \approx 50, \quad m \approx 1700 \quad (\text{over } \mathbb{F}_{2^8})$$

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The magic

1. Fix number of iterations.
2. For *large m*, message V_i uniformly random conditional on dimension.
3. For *large l*, output dimension determined by input dimension.
4. For $\omega = 1/k$, only dimension $l\omega$ or 0 is possible.

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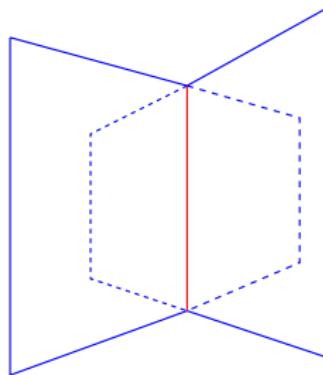
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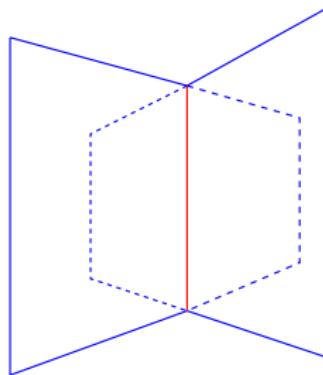
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“For large ℓ , output dimension determined by input dimension”



$$d_{1\cap 2} \approx \max(d_1 + d_2 - d_{\text{amb}}, 0).$$

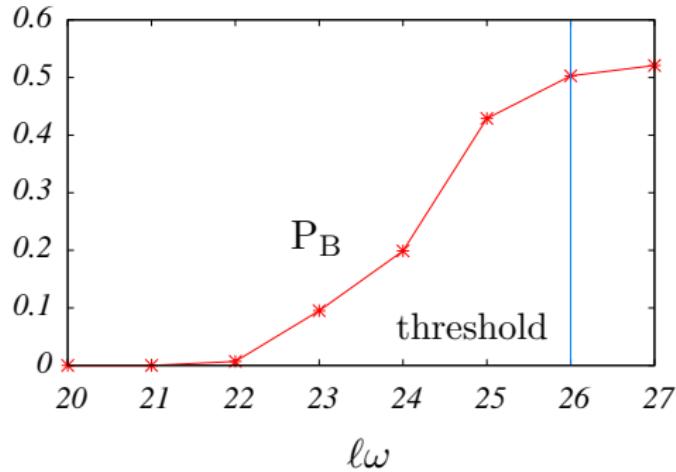
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Simulations at small packet sizes

$m = 42, \ell = 136$ (*LP optimized degree sequence*)



Conclusion 1

q -ary LDPC's (Davey, MacKay/ Burshtein, Miller/ Rathi, Urbanke)
Mixed outcomes

Exact density evolution for large q

Capacity achieving ensembles

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Capacity achieving ensembles

Conclusion 2

LDPC codes achieve capacity (under message passing decoding)
over the erasure channel (Luby et al. 97) ...

... and the rank metric channel

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